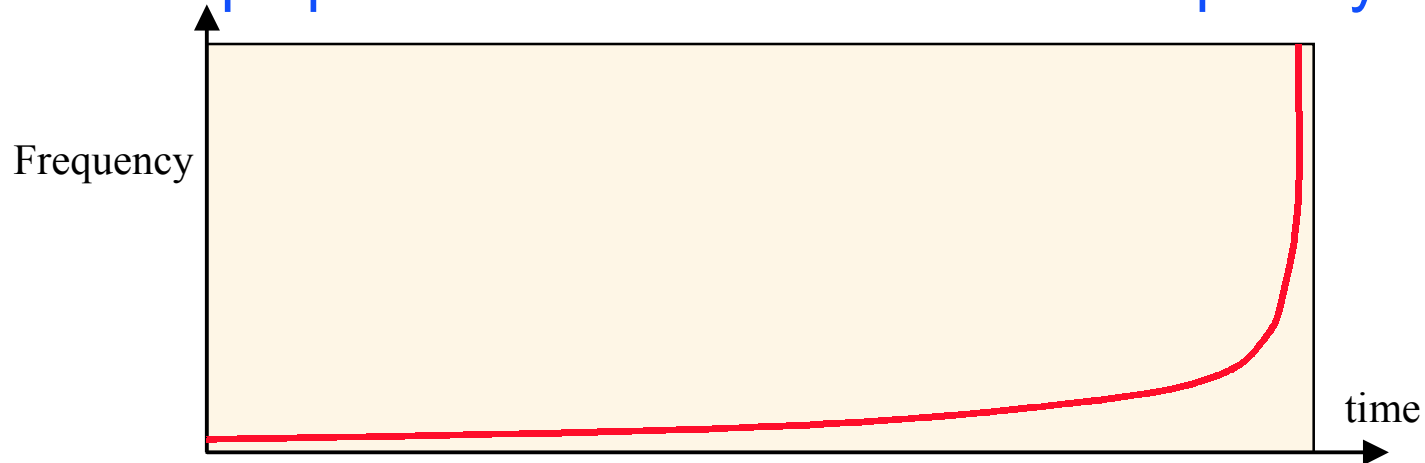


# Multi-Band Template Analysis for inspiraling binaries

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Caltech & LAPP-Annecy

April 2001, NSF Review

- Binary search requires large computing resources, especially :
  - » for low mass
  - » if we start at low frequency
- The chirp spend most of its time at low frequency.



- » The number of templates depends of the chirp length and the maximum chirp frequency
- How to reduce the computing cost?

# LIGO The Multi Band Template Analysis

- Principle: Split the analysis in a few frequency bands

- » Equivalent to transform a single detector to a network of detectors.

$$LLR(t, M) = \int_{f_{\min}}^{f_{\max}} h(f)T(M, f)df = \int_{f_{\min}}^{f_1} h(f)T(M, f)df + \int_{f_1}^{f_{\max}} h(f)T(M, f)df$$

- » Analyze independently each band

- » Combine coherently the analysis result like for a network of detectors

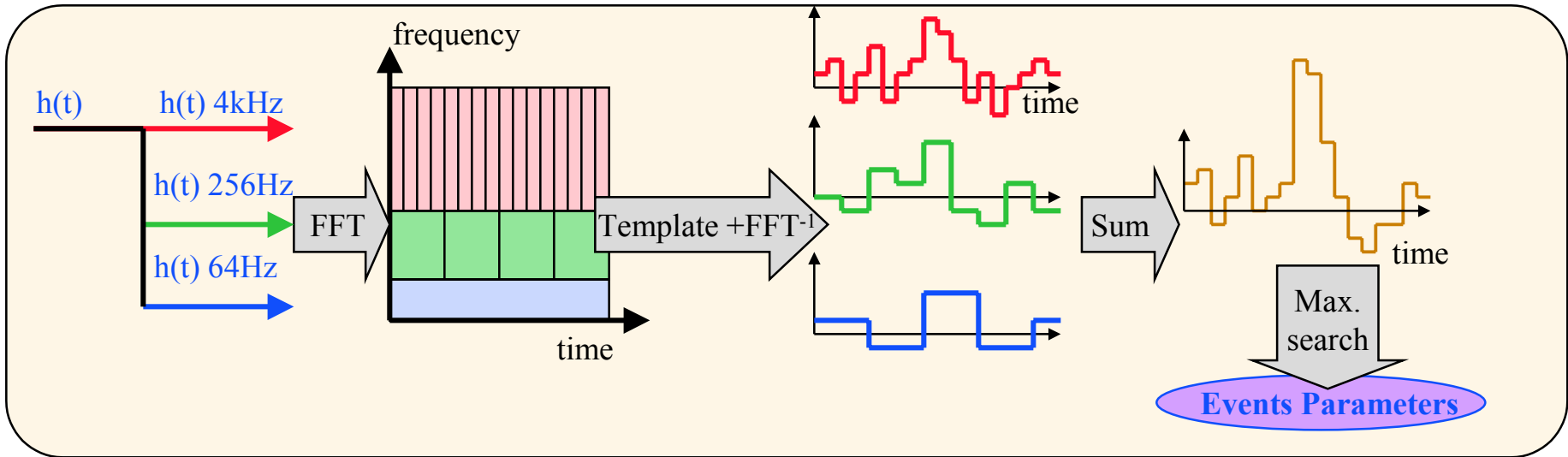
- Each analysis is cheaper because

- » The number of templates is reduce

- » The FFT are shorter

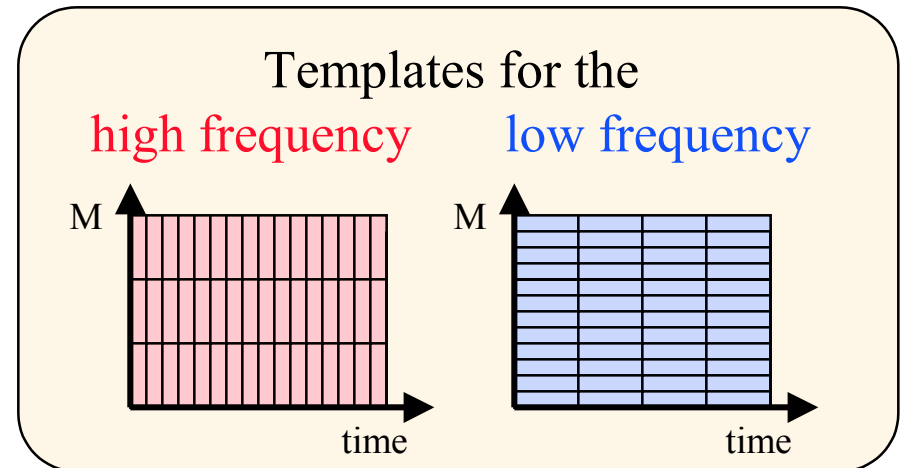
- Remarks:

- » The SNR should be unchanged.

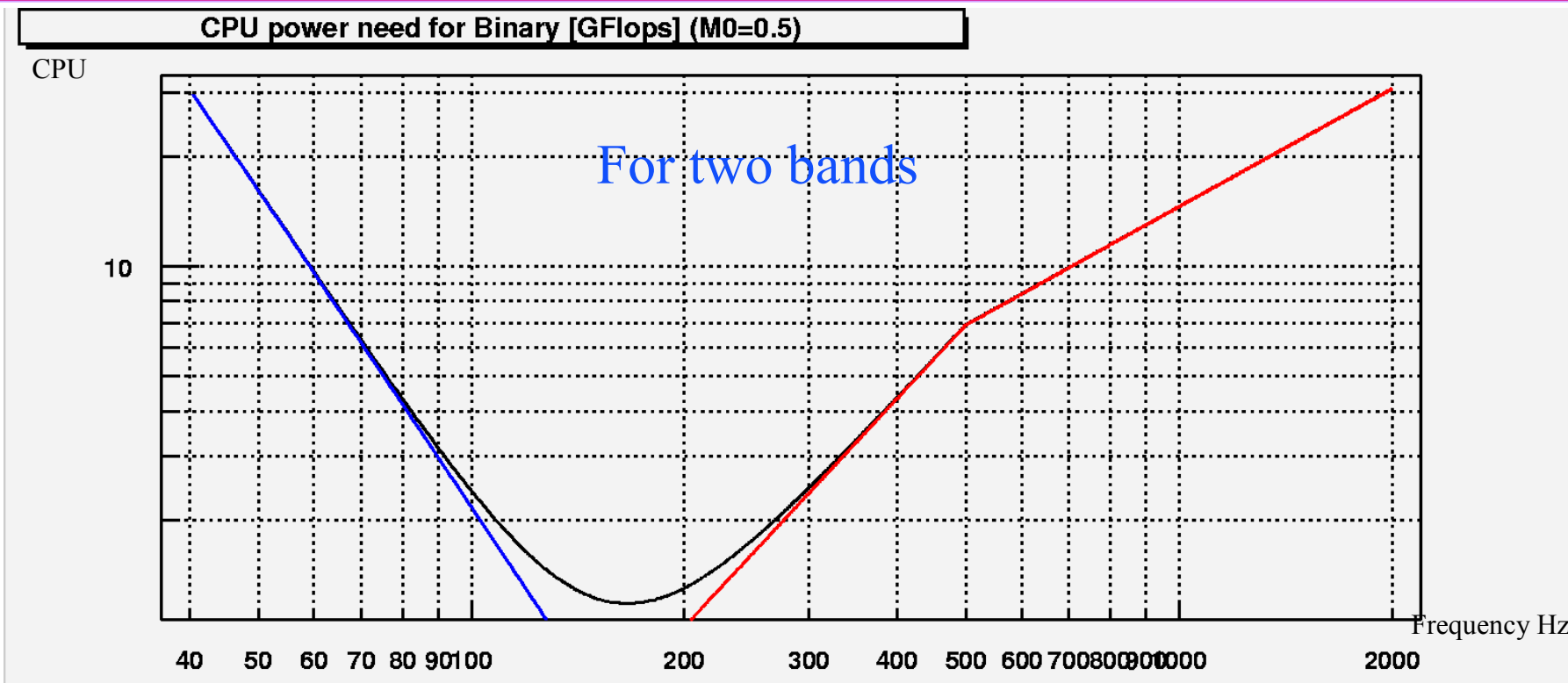


## ● Remarks:

- » All sub analysis cover the same parameter space **BUT** may have different grids.
- » Need interpolation to combine the results and search for the maximum.
- » All FFT are small FFT.



# LIGO Estimation of CPU resources



$$\text{CPU} = K f_{\min}^{-8/3} f_s \log_2(f_{\min} f_{\max}) \quad (\text{simplify model: } N_{\text{template}} \cdot \text{FFT cost})$$

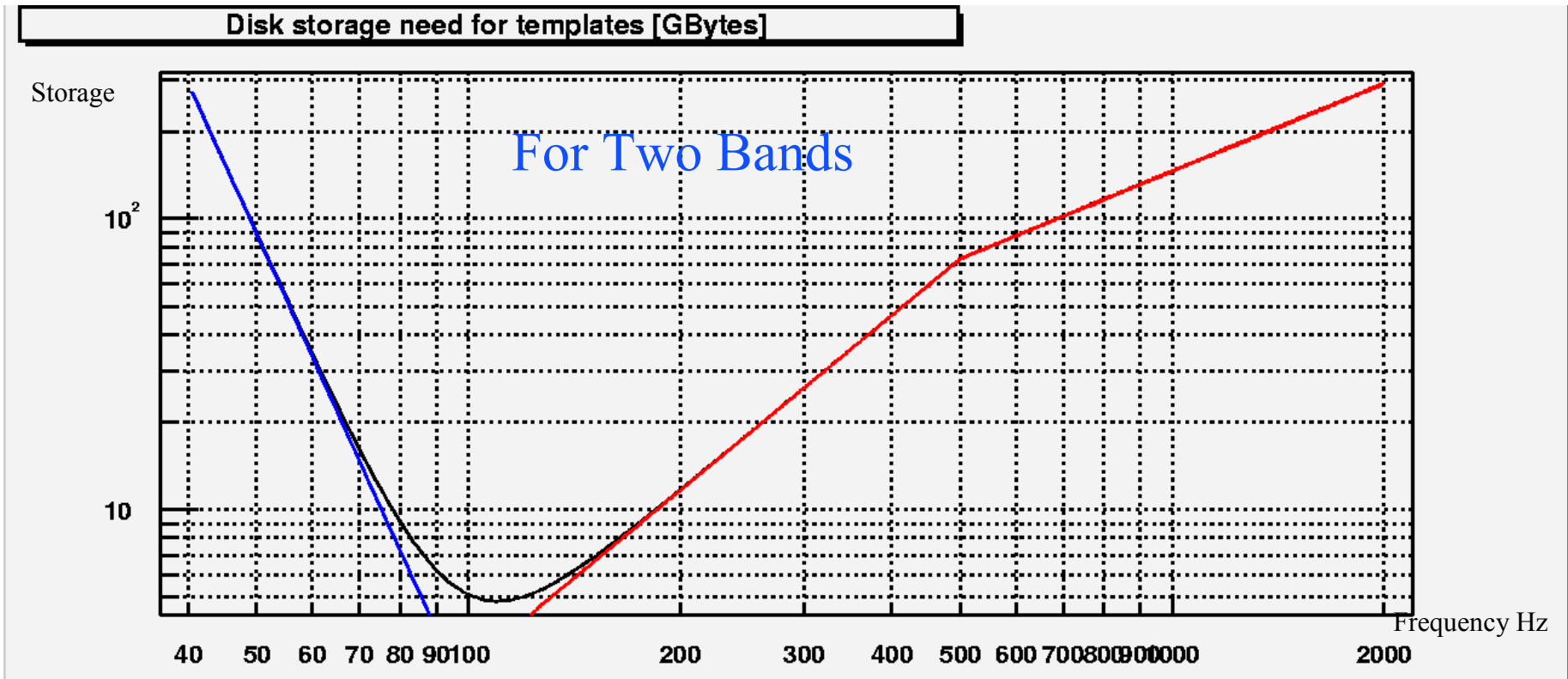
$$T = \text{Template length (seconds)} = T_0 f_{\min}^{-8/3}$$

$$N_{\text{template}} = T / \text{template spacing}$$

$$N_{\text{sample}} = 2T f_{\max}$$

$$\text{CPU} = N_{\text{template}} 6N_{\text{sample}} \log_2(N_{\text{sample}}) / T = K f_{\min}^{-8/3} f_{\max} \log_2(f_{\min} f_{\max})$$

# LIGO Estimation Template Storage



$$\text{Storage} = K f_{\min}^{-16/3} f_s \log_2(f_{\min} f_{\max}) \quad (\text{simplify model: } N_{\text{template}} \cdot \text{tempSize})$$

$$T = \text{Template length (seconds)}. = T_0 f_{\min}^{-8/3}$$

$$N_{\text{template}} = T / \text{template spacing}$$

$$N_{\text{sample}} = 2T f_{\max}$$

$$\text{Storage} = 2 N_{\text{template}} N_{\text{sample}} = K f_{\min}^{-16/3} f_{\max} \log_2(f_{\min} f_{\max})$$

# LIGO Estimation of computing resources

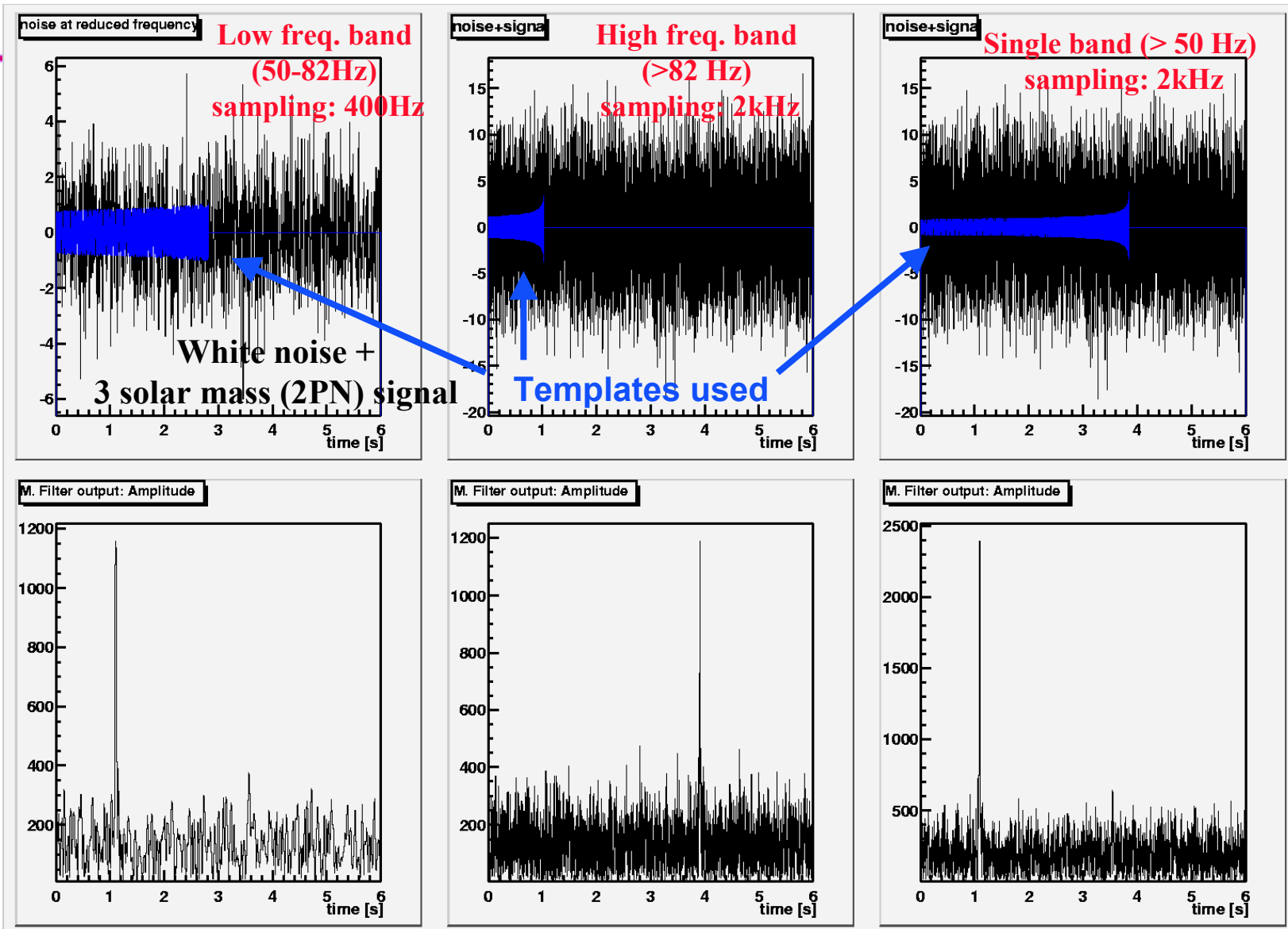
- If  $f_{\min} = 40$  Hz,  $f_{\max} = 2$  kHz,  $M_{\min} = 0.5$  M

	1 Band	2 Bands	3 Bands
CPU(Gflops)	30	1.3	0.6
Storage (Gbytes)	300	5	2.4
T. size (Mbytes)	2	0.13	0.04

- If  $f_{\min} = 20$  Hz,  $f_{\max} = 2$  kHz,  $M_{\min} = 0.5$  M

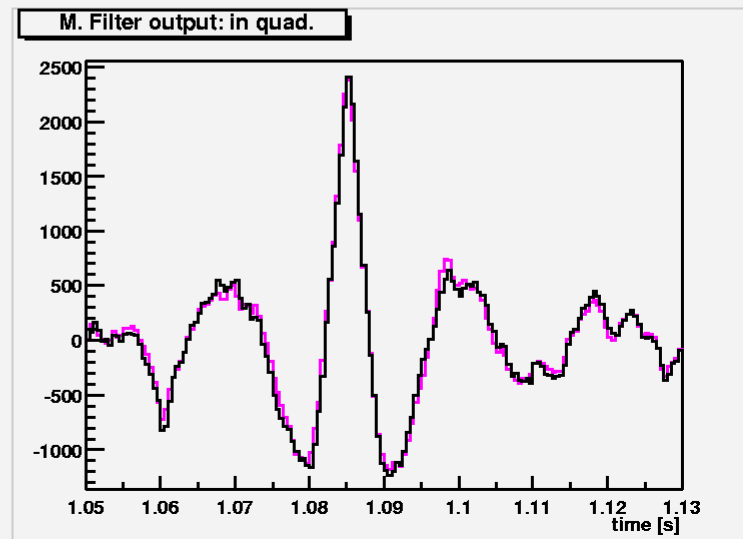
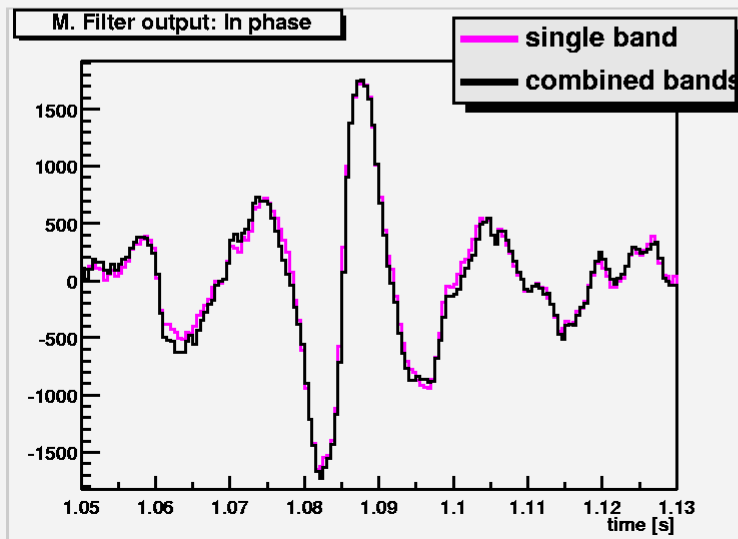
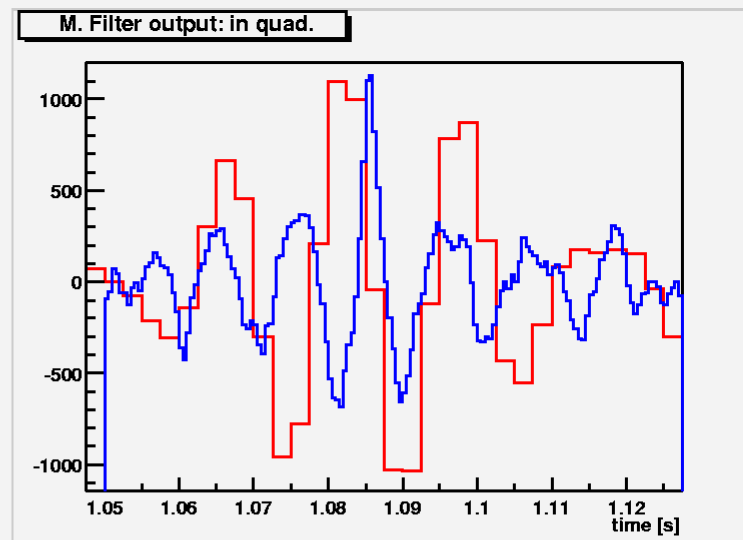
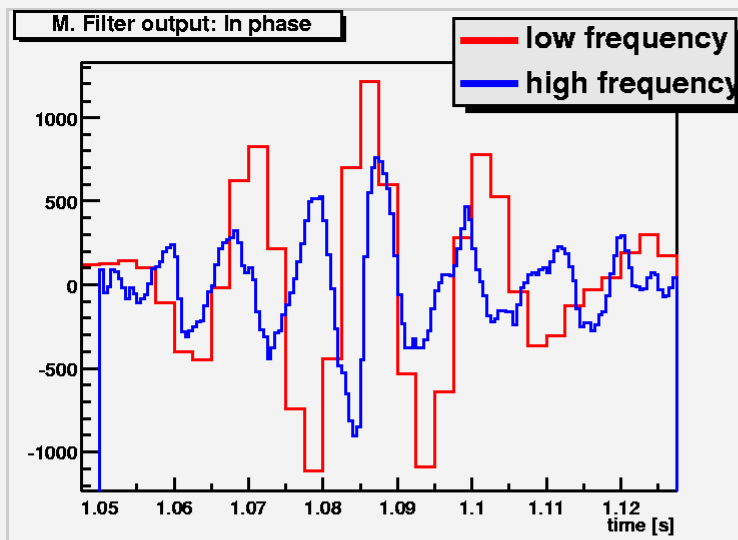
	1 Band	2 Bands	3 Bands
CPU(Gflops)	200	4.3	1.3
Storage (Gbytes)	10000	100	43
T. size (Mbytes)	11	0.6	0.2

## Does it work? Test with 2 bands

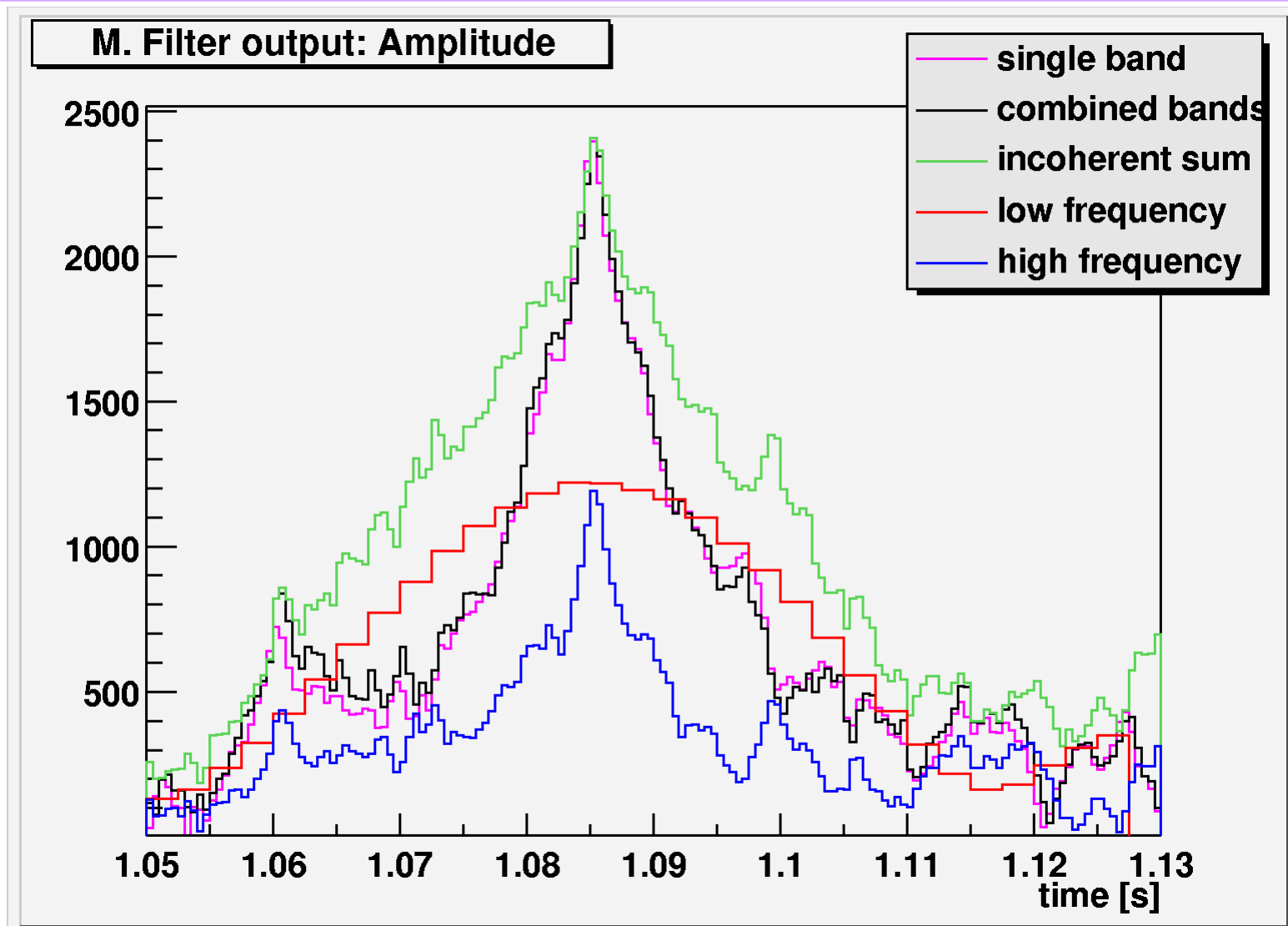




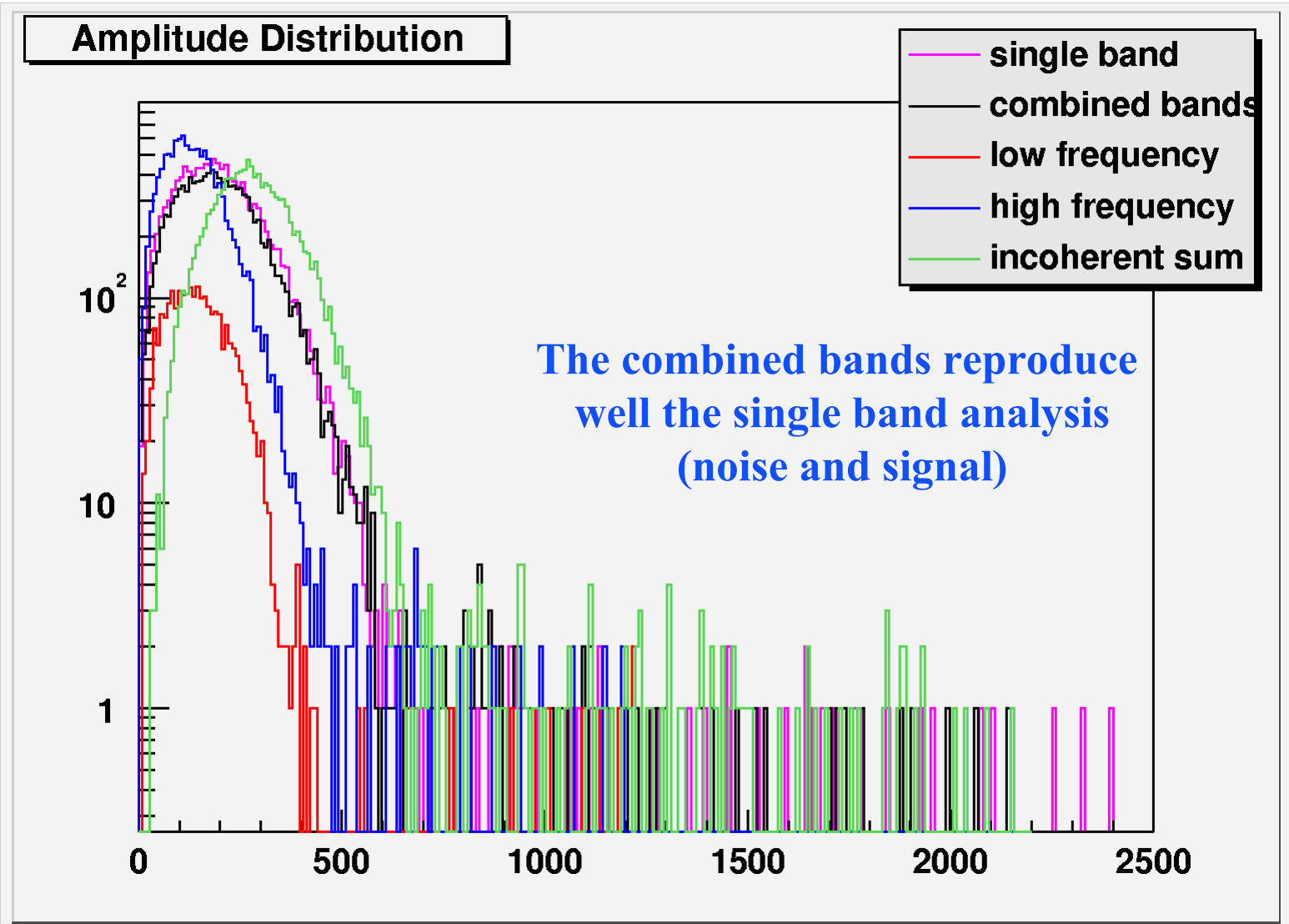
# Zoom on each templates



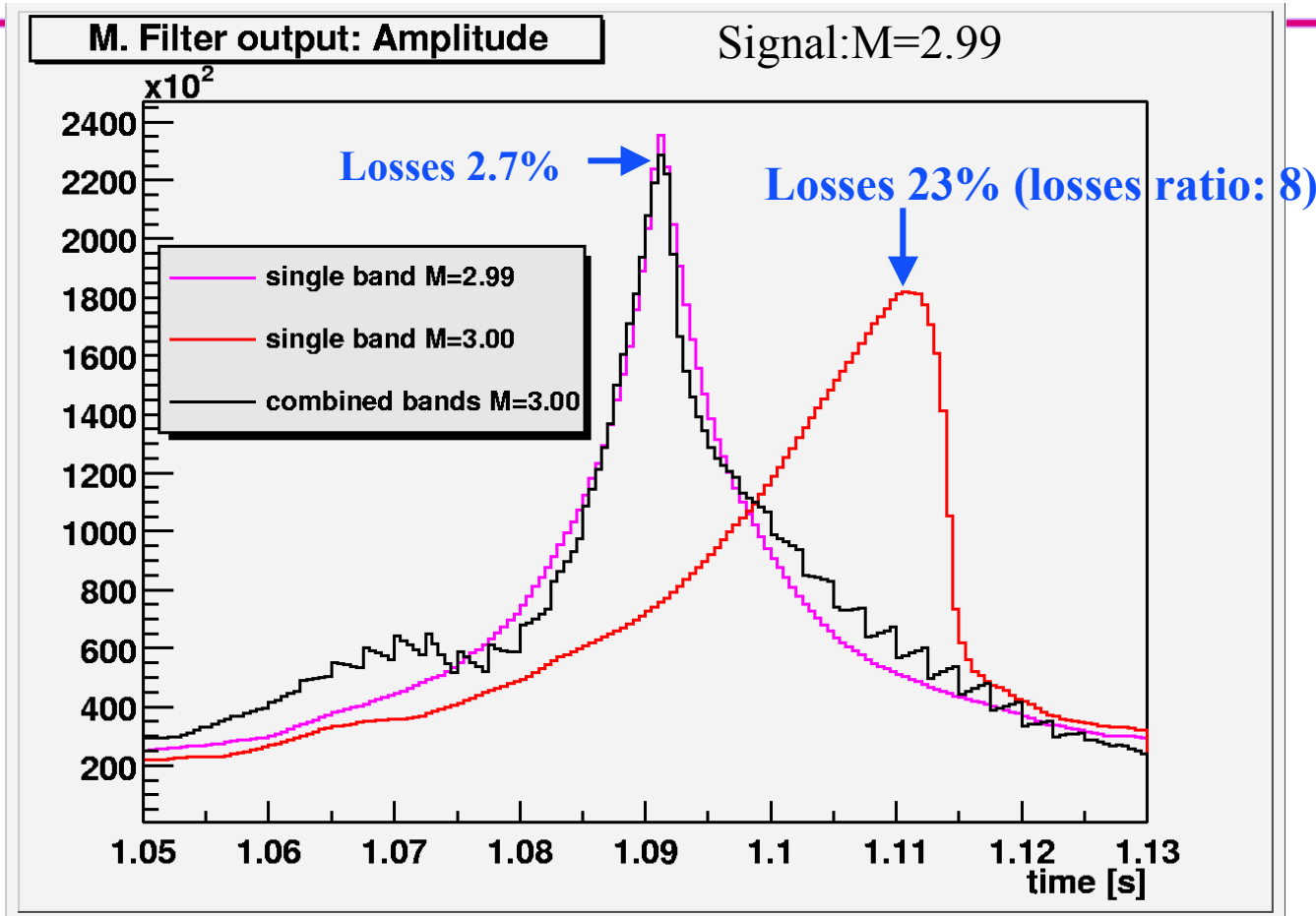
# Comparison of the outputs



# Comparison of the noises



# LIGO Signal and Template mismatched



The parameter space is 8 time larger,

The templates are 5 times smaller

⇒ reduction of ~10 for CPU, 40 for template storage

- The Multi Band Template Analysis has many advantages
  - » No SNR change
  - » Reduce the computing requirements
  - » Work on small FFT (fit in the CPU cache, use single precision)
  - » Build-in hierarchical approach without compromise on SNR
  - » Build-in consistency tests
  
- More study in progress
  - » Implementation problems? Is the gain as good as expected?
    - ⇒ Building a prototype code